# Foundation of Cryptography, Lecture 3
# Hardcore Predicates for Any One-way Function

Iftach Haitner, Tel Aviv University

Tel Aviv University.

April 9, 2013

# Hardcore Predicates

**Definition 1 (hardcore predicates)**

A poly-time computable $b \colon \{0,1\}^n \mapsto \{0,1\}$ is an hardcore predicate of $f \colon \{0,1\}^n \mapsto \{0,1\}^n$, if

$$\Pr_{x \leftarrow \{0,1\}^n}[\mathsf{P}(f(x)) = b(x)] \leq \frac{1}{2} + \mathrm{neg}(n)$$

for any PPT $\mathsf{P}$.

# Hardcore Predicates

**Definition 1 (hardcore predicates)**

A poly-time computable $b\colon \{0,1\}^n \mapsto \{0,1\}$ is an hardcore predicate of $f\colon \{0,1\}^n \mapsto \{0,1\}^n$, if

$$\Pr_{x \leftarrow \{0,1\}^n}[\mathsf{P}(f(x)) = b(x)] \leq \frac{1}{2} + \text{neg}(n)$$

for any PPT $\mathsf{P}$.

- (precious class) an hardcore predicate for a permutation $\implies$ PRG

## Hardcore Predicates

### Definition 1 (hardcore predicates)

A poly-time computable $b\colon \{0,1\}^n \mapsto \{0,1\}$ is an hardcore predicate of $f\colon \{0,1\}^n \mapsto \{0,1\}^n$, if

$$\Pr_{x \leftarrow \{0,1\}^n}[\mathsf{P}(f(x)) = b(x)] \leq \frac{1}{2} + \mathsf{neg}(n)$$

for any PPT $\mathsf{P}$.

- (precious class) an hardcore predicate for a permutation $\implies$ PRG
- Can there exist a “generic" hardcore predicate?

# Weak hardcore predicate

### Theorem 2 (Proven in HW)

*Let* $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ *be a OWF, and define* $g\colon \{0,1\}^n \times [n] \mapsto \{0,1\}^n \times [n]$ *as* $g(x,i) = f(x), i$ *and* $b(x,i) = x[i]$*. Then*

$$\Pr_{x \leftarrow \{0,1\}^n, i \leftarrow [n]}[\mathsf{A}(f(x),i) = x[i]] \le 1 - 1/2n$$

*for any* PPT $\mathsf{A}$*.*

# Weak hardcore predicate

**Theorem 2 (Proven in HW)**

*Let* $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ *be a OWF, and define* $g\colon \{0,1\}^n \times [n] \mapsto \{0,1\}^n \times [n]$ *as* $g(x,i) = f(x), i$ *and* $b(x,i) = x[i]$. *Then*

$$\Pr_{x \leftarrow \{0,1\}^n, i \leftarrow [n]}[\mathsf{A}(f(x),i) = x[i]] \leq 1 - 1/2n$$

*for any* PPT A.

- $b$ is a “weak" hardcore predicate of $g$

## Weak hardcore predicate

**Theorem 2 (Proven in HW)**

*Let* $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ *be a OWF, and define* $g\colon \{0,1\}^n \times [n] \mapsto \{0,1\}^n \times [n]$ *as* $g(x,i) = f(x), i$ *and* $b(x,i) = x[i]$. *Then*

$$\Pr_{x \leftarrow \{0,1\}^n, i \leftarrow [n]}[\mathsf{A}(f(x), i) = x[i]] \leq 1 - 1/2n$$

*for any* PPT $\mathsf{A}$.

- $b$ is a “weak" hardcore predicate of $g$
- $b$ can be “amplified" to a (strong) hardcore predicate

# The Goldreich-Levin Hadrcore predicate

**Theorem 3 (Goldreich-Levin)**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x, r\rangle_2$ is an hardcore predicate of $g$.*

# The Goldreich-Levin Hadrcore predicate

**Theorem 3 (Goldreich-Levin)**

*Let* $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ *be a OWF, and define* $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ *as* $g(x,r) = f(x), r$. *Then* $b(x,r) = \langle x, r \rangle_2$ *is an hardcore predicate of* $g$.

- $\langle x, r \rangle_2 := (\sum_{i=1}^{n} x_i \cdot r_i) \bmod 2$.

# The Goldreich-Levin Hadrcore predicate

**Theorem 3 (Goldreich-Levin)**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x,r \rangle_2$ is an hardcore predicate of $g$.*

- $\langle x,r \rangle_2 := (\sum_{i=1}^{n} x_i \cdot r_i) \bmod 2$.
- Note that if $f$ is one-to-one, then so is $g$.

Section 1

# Proving GL, The Information Theoretic Case

## GL is hard for regular functions

### Definition 4 (min-entropy)

The min entropy of a random variable $X$, is defined as

$$\mathrm{H}_\infty(X) := \min_{y \in \mathrm{Supp}(X)} \log \frac{1}{\Pr_X[y]}.$$

# GL is hard for regular functions

**Definition 4 (min-entropy)**

The min entropy of a random variable $X$, is defined as

$$\mathrm{H}_\infty(X) := \min_{y \in \mathrm{Supp}(X)} \log \frac{1}{\Pr_X[y]}.$$

Examples:

- $X$ is uniform over a set of size $2^k$, then $\mathrm{H}_\infty(X) = k$.

## GL is hard for regular functions

**Definition 4 (min-entropy)**

The min entropy of a random variable $X$, is defined as

$$\mathrm{H}_\infty(X) := \min_{y \in \mathrm{Supp}(X)} \log \frac{1}{\Pr_X[y]}.$$

Examples:

- $X$ is uniform over a set of size $2^k$, then $\mathrm{H}_\infty(X) = k$.
- $(X \mid f(X) = y)$, where $f \colon \{0,1\}^n \mapsto \{0,1\}^n$ is $2^k$ to $1$ and $X$ is uniform over $\{0,1\}^n$

# Pairwise independent hashing

**Definition 5 (pairwise independent hash functions)**

A function family $\mathcal{H}$ from $\{0,1\}^n$ to $\{0,1\}^m$ is pairwise independent, if for every $x \neq x' \in \{0,1\}^n$ and $y, y' \in \{0,1\}^m$, it holds that $\Pr_{h \leftarrow \mathcal{H}}[h(x) = y \wedge h(x') = y')] = 2^{-2m}$.

# Pairwise independent hashing

**Definition 5 (pairwise independent hash functions)**

A function family $\mathcal{H}$ from $\{0,1\}^n$ to $\{0,1\}^m$ is pairwise independent, if for every $x \neq x' \in \{0,1\}^n$ and $y, y' \in \{0,1\}^m$, it holds that $\Pr_{h \leftarrow \mathcal{H}}[h(x) = y \wedge h(x') = y')] = 2^{-2m}$.

**Lemma 6 (leftover hash lemma)**

*Let $X$ be a random variable over $\{0,1\}^n$ with $\mathrm{H}_\infty(X) \geq k$ and let $\mathcal{H}$ be a family of pairwise independent hash functions from $\{0,1\}^n$ to $\{0,1\}^m$, then*

$$\mathrm{SD}((H, H(X)), (H, U_m)) \leq 2^{(m-k-2))/2},$$

*where $H$ is uniformly distributed over $\mathcal{H}$, and $U_m$ over $\{0,1\}^m$.*

## Efficient function families

**Definition 7 (efficient function families)**

An ensemble of function families $\mathcal{F} = \{\mathcal{F}_n\}_{n\in\mathbb{N}}$ is efficient, if

**Samplable.** $\mathcal{F}$ is samplable in polynomial-time: there exists a PPT that given $1^n$, outputs (the description of) a uniform element in $\mathcal{F}_n$.

**Efficient.** There exists a polynomial-time algorithm that given $x \in \{0,1\}^n$ and (a description of) $f \in \mathcal{F}_n$, outputs $f(x)$.

## Hardcore predicate for regular functions

**Lemma 8**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a $d(n) \in 2^{\omega(\log n)}$ regular function and let $\mathcal{H} = \{\mathcal{H}_n\}$ be an efficient family of Boolean pairwise independent hash functions over $\{0,1\}^n$. Define $g\colon \{0,1\}^n \times \mathcal{H}_n \mapsto \{0,1\}^n \times \mathcal{H}_n$ as*

$$g(x,h) = (f(x),h),$$

*then $b(x,h) = h(x)$ is an hardcore predicate of $g$.*

# Hardcore predicate for regular functions

## Lemma 8

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a $d(n) \in 2^{\omega(\log n)}$ regular function and let $\mathcal{H} = \{\mathcal{H}_n\}$ be an efficient family of Boolean pairwise independent hash functions over $\{0,1\}^n$. Define $g\colon \{0,1\}^n \times \mathcal{H}_n \mapsto \{0,1\}^n \times \mathcal{H}_n$ as*

$$g(x,h) = (f(x),h),$$

*then $b(x,h) = h(x)$ is an hardcore predicate of $g$.*

How does it relate to Goldreich-Levin?

# Hardcore predicate for regular functions

## Lemma 8

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a $d(n) \in 2^{\omega(\log n)}$ regular function and let $\mathcal{H} = \{\mathcal{H}_n\}$ be an efficient family of Boolean pairwise independent hash functions over $\{0,1\}^n$. Define $g\colon \{0,1\}^n \times \mathcal{H}_n \mapsto \{0,1\}^n \times \mathcal{H}_n$ as*

$$g(x,h) = (f(x), h),$$

*then $b(x,h) = h(x)$ is an hardcore predicate of $g$.*

How does it relate to Goldreich-Levin? $b(x,r)$ is (almost) a family of pairwise independent hash functions.

# Hardcore predicate for regular functions

## Lemma 8

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a $d(n) \in 2^{\omega(\log n)}$ regular function and let $\mathcal{H} = \{\mathcal{H}_n\}$ be an efficient family of Boolean pairwise independent hash functions over $\{0,1\}^n$. Define $g\colon \{0,1\}^n \times \mathcal{H}_n \mapsto \{0,1\}^n \times \mathcal{H}_n$ as*

$$g(x,h) = (f(x),h),$$

*then $b(x,h) = h(x)$ is an hardcore predicate of $g$.*

How does it relate to Goldreich-Levin? $b(x,r)$ is (almost) a family of pairwise independent hash functions.

We prove Lemma 8 by showing that

## Claim 9

$\mathrm{SD}\left((f(U_n),H,H(U_n)),(f(U_n),H,U_1)\right) = \mathrm{neg}(n)$, where $H = H_n$ is uniformly distributed over $\mathcal{H}_n$.

## Hardcore predicate for regular functions

**Lemma 8**

*Let* $f: \{0,1\}^n \mapsto \{0,1\}^n$ *be a* $d(n) \in 2^{\omega(\log n)}$ *regular function and let* $\mathcal{H} = \{\mathcal{H}_n\}$ *be an efficient family of Boolean pairwise independent hash functions over* $\{0,1\}^n$*. Define* $g: \{0,1\}^n \times \mathcal{H}_n \mapsto \{0,1\}^n \times \mathcal{H}_n$ *as*

$$g(x,h) = (f(x),h),$$

*then* $b(x,h) = h(x)$ *is an hardcore predicate of* $g$*.*

How does it relate to Goldreich-Levin? $b(x,r)$ is (almost) a family of pairwise independent hash functions.

We prove Lemma 8 by showing that

**Claim 9**

$\mathrm{SD}\left((f(U_n), H, H(U_n)), (f(U_n), H, U_1)\right) = \mathrm{neg}(n)$, where $H = H_n$ is uniformly distributed over $\mathcal{H}_n$.

Does this conclude the proof?

## Proving Claim 9

Proof: For $y \in f(\{0,1\}^n) := \{f(x) \colon x \in \{0,1\}^n\}$, let $X_y$ be uniformly distributed over $f^{-1}(y) := \{x \in \{0,1\}^n \colon f(x) = y\}$.

# Proving Claim 9

Proof: For $y \in f(\{0,1\}^n) := \{f(x) \colon x \in \{0,1\}^n\}$, let $X_y$ be uniformly distributed over $f^{-1}(y) := \{x \in \{0,1\}^n \colon f(x) = y\}$.

$$\begin{aligned}
&\mathrm{SD}((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathrm{SD}\big((f(U_n), H, H(U_n) \mid f(U_n) = y) \\
&\qquad\qquad\qquad\qquad , (f(U_n), H, U_1 \mid f(U_n) = y)\big)
\end{aligned}$$

# Proving Claim 9

Proof: For $y \in f(\{0,1\}^n) := \{f(x) \colon x \in \{0,1\}^n\}$, let $X_y$ be uniformly distributed over $f^{-1}(y) := \{x \in \{0,1\}^n \colon f(x) = y\}$.

$$
\begin{aligned}
&\mathrm{SD}((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathrm{SD}\big((f(U_n), H, H(U_n) \mid f(U_n) = y) \\
&\qquad\qquad\qquad\qquad\qquad\qquad , (f(U_n), H, U_1 \mid f(U_n) = y)\big) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathrm{SD}\left((y, H, H(X_y)), (y, H, U_1)\right)
\end{aligned}
$$

## Proving Claim 9

Proof: For $y \in f(\{0,1\}^n) := \{f(x) \colon x \in \{0,1\}^n\}$, let $X_y$ be uniformly distributed over $f^{-1}(y) := \{x \in \{0,1\}^n \colon f(x) = y\}$.

$$
\begin{aligned}
&\mathrm{SD}((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathrm{SD}\big((f(U_n), H, H(U_n) \mid f(U_n) = y) \\
&\qquad\qquad , (f(U_n), H, U_1 \mid f(U_n) = y)\big) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathrm{SD}\left((y, H, H(X_y)), (y, H, U_1)\right) \\
&\leq \max_{y \in f(\{0,1\}^n)} \mathrm{SD}((y, H, H(X_y)), (y, H, U_1))
\end{aligned}
$$

# Proving Claim 9

Proof: For $y \in f(\{0,1\}^n) := \{f(x) \colon x \in \{0,1\}^n\}$, let $X_y$ be uniformly distributed over $f^{-1}(y) := \{x \in \{0,1\}^n \colon f(x) = y\}$.

$$
\begin{aligned}
& \mathsf{SD}((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathsf{SD}\big((f(U_n), H, H(U_n) \mid f(U_n) = y) \\
&\qquad\qquad , (f(U_n), H, U_1 \mid f(U_n) = y)\big) \\
&= \sum_{y \in f(\{0,1\}^n)} \Pr[f(U_n) = y] \cdot \mathsf{SD}\left((y, H, H(X_y)), (y, H, U_1)\right) \\
&\leq \max_{y \in f(\{0,1\}^n)} \mathsf{SD}((y, H, H(X_y)), (y, H, U_1)) \\
&= \max_{y \in f(\{0,1\}^n)} \mathsf{SD}((H, H(X_y)), (H, U_1))
\end{aligned}
$$

Since $\mathrm{H}_{\infty}(X_y) = \log(d(n))$ for any $y \in f(\{0,1\}^n)$,

Since $\mathrm{H}_\infty(X_y) = \log(d(n))$ for any $y \in f(\{0,1\}^n)$,
The leftover hash lemma (Lemma 6) yields that

$$\begin{aligned} \mathrm{SD}((H, H(X_y)), (H, U_1)) &\leq 2^{(1-\mathrm{H}_\infty(X_y)-2))/2} \\ &= 2^{(1-\log(d(n)))/2} = \mathrm{neg}(n). \quad \square \end{aligned}$$

# Further remarks

**Remark 10**

- Can output $\Theta(\log d(n))$ bits

# Further remarks

**Remark 10**

- Can output $\Theta(\log d(n))$ bits
- $g$ and $b$ have partial domains.

Section 2

# Proving GL, The Computational Case

## Proving Goldreich-Levin Theorem

Recall

**Theorem 11 (Goldreich-Levin)**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x, r\rangle_2$ is an hardcore predicate of $g$.*

## Proving Goldreich-Levin Theorem

Recall

**Theorem 11 (Goldreich-Levin)**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x, r\rangle_2$ is an hardcore predicate of $g$.*

Proof: Assume $\exists$ PPT A, $p \in$ poly and infinite set $\mathcal{I} \subseteq \mathbb{N}$ with

$$\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] \geq \frac{1}{2} + \frac{1}{p(n)}, \tag{1}$$

for any $n \in \mathcal{I}$, where $U_n$ and $R_n$ are uniformly (and independently) distributed over $\{0,1\}^n$.

## Proving Goldreich-Levin Theorem

Recall

**Theorem 11 (Goldreich-Levin)**

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x, r\rangle_2$ is an hardcore predicate of $g$.*

Proof: Assume $\exists$ PPT A, $p \in$ poly and infinite set $\mathcal{I} \subseteq \mathbb{N}$ with

$$\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] \geq \frac{1}{2} + \frac{1}{p(n)}, \tag{1}$$

for any $n \in \mathcal{I}$, where $U_n$ and $R_n$ are uniformly (and independently) distributed over $\{0,1\}^n$.

We show $\exists$ PPT B and $q \in$ poly with

$$\Pr_{y \leftarrow f(U_n)}[\mathsf{B}(y) \in f^{-1}(y)] \geq \frac{1}{q(n)}, \tag{2}$$

for every $n \in \mathcal{I}$.

## Proving Goldreich-Levin Theorem

Recall

### Theorem 11 (Goldreich-Levin)

*Let $f\colon \{0,1\}^n \mapsto \{0,1\}^n$ be a OWF, and define $g\colon \{0,1\}^n \times \{0,1\}^n \mapsto \{0,1\}^n \times \{0,1\}^n$ as $g(x,r) = f(x), r$. Then $b(x,r) = \langle x, r\rangle_2$ is an hardcore predicate of $g$.*

Proof: Assume $\exists$ PPT A, $p \in$ poly and infinite set $\mathcal{I} \subseteq \mathbb{N}$ with

$$\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] \geq \frac{1}{2} + \frac{1}{p(n)}, \tag{1}$$

for any $n \in \mathcal{I}$, where $U_n$ and $R_n$ are uniformly (and independently) distributed over $\{0,1\}^n$.

We show $\exists$ PPT B and $q \in$ poly with

$$\Pr_{y \leftarrow f(U_n)}[\mathsf{B}(y) \in f^{-1}(y)] \geq \frac{1}{q(n)}, \tag{2}$$

for every $n \in \mathcal{I}$. In the following fix $n \in \mathcal{I}$.

## Focusing on a good set

### Claim 12

There exists a set $\mathcal{S} \subseteq \{0,1\}^n$ with

1. $\frac{|\mathcal{S}|}{2^n} \geq \frac{1}{2p(n)}$, and
2. $\alpha(x) := \Pr[\mathsf{A}(f(x), R_n) = b(x, R_n)] \geq \frac{1}{2} + \frac{1}{2p(n)}$, $\forall x \in \mathcal{S}$.

## Focusing on a good set

### Claim 12

There exists a set $\mathcal{S} \subseteq \{0,1\}^n$ with

1. $\frac{|\mathcal{S}|}{2^n} \geq \frac{1}{2p(n)}$, and
2. $\alpha(x) := \Pr[\mathsf{A}(f(x), R_n) = b(x, R_n)] \geq \frac{1}{2} + \frac{1}{2p(n)}$, $\forall x \in S$.

Proof: Let $\mathcal{S} := \{x \in \{0,1\}^n \colon \alpha(x) \geq \frac{1}{2} + \frac{1}{2p(n)}\}$. It follows that

$$\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] \quad \leq \Pr[U_n \notin \mathcal{S}] \cdot \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}]$$

## Focusing on a good set

### Claim 12

There exists a set $\mathcal{S} \subseteq \{0,1\}^n$ with

1. $\frac{|\mathcal{S}|}{2^n} \geq \frac{1}{2p(n)}$, and
2. $\alpha(x) := \Pr[\mathsf{A}(f(x), R_n) = b(x, R_n)] \geq \frac{1}{2} + \frac{1}{2p(n)}, \forall x \in S$.

Proof: Let $\mathcal{S} := \{x \in \{0,1\}^n \colon \alpha(x) \geq \frac{1}{2} + \frac{1}{2p(n)}\}$. It follows that

$$\begin{aligned}\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] &\leq \Pr[U_n \notin \mathcal{S}] \cdot \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \\ &\leq \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \square\end{aligned}$$

# Focusing on a good set

## Claim 12

There exists a set $\mathcal{S} \subseteq \{0,1\}^n$ with

1. $\frac{|\mathcal{S}|}{2^n} \geq \frac{1}{2p(n)}$, and
2. $\alpha(x) := \Pr[\mathsf{A}(f(x), R_n) = b(x, R_n)] \geq \frac{1}{2} + \frac{1}{2p(n)}$, $\forall x \in S$.

Proof: Let $\mathcal{S} := \{x \in \{0,1\}^n \colon \alpha(x) \geq \frac{1}{2} + \frac{1}{2p(n)}\}$. It follows that

$$\begin{aligned} \Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] &\leq \Pr[U_n \notin \mathcal{S}] \cdot \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \\ &\leq \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \square \end{aligned}$$

We will present $q \in \text{poly}$ and a PPT $B$ such that

$$\Pr[\mathsf{B}(y = f(x)) \in f^{-1}(y) \geq \frac{1}{q(n)}, \tag{3}$$

for every $x \in \mathcal{S}$.

# Focusing on a good set

## Claim 12

There exists a set $\mathcal{S} \subseteq \{0,1\}^n$ with

1. $\frac{|\mathcal{S}|}{2^n} \geq \frac{1}{2p(n)}$, and
2. $\alpha(x) := \Pr[\mathsf{A}(f(x), R_n) = b(x, R_n)] \geq \frac{1}{2} + \frac{1}{2p(n)}$, $\forall x \in S$.

Proof: Let $\mathcal{S} := \{x \in \{0,1\}^n \colon \alpha(x) \geq \frac{1}{2} + \frac{1}{2p(n)}\}$. It follows that

$$\begin{aligned}\Pr[\mathsf{A}(g(U_n, R_n)) = b(U_n, R_n)] &\leq \Pr[U_n \notin \mathcal{S}] \cdot \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \\ &\leq \left(\frac{1}{2} + \frac{1}{2p(n)}\right) + \Pr[U_n \in \mathcal{S}] \square\end{aligned}$$

We will present $q \in \text{poly}$ and a PPT $B$ such that

$$\Pr[\mathsf{B}(y = f(x)) \in f^{-1}(y) \geq \frac{1}{q(n)}, \tag{3}$$

for every $x \in \mathcal{S}$. In the following we fix $x \in \mathcal{S}$.

## The perfect case $\alpha(x) = 1$

- For every $i \in [n]$ it holds that

$$A(f(x), e^i) = b(x, e^i),$$

where $e^i = (\underbrace{0, \ldots, 0}_{i-1}, 1, \underbrace{0, \ldots, 0}_{n-i})$.

## The perfect case $\alpha(x) = 1$

- For every $i \in [n]$ it holds that

$$\mathsf{A}(f(x), e^i) = b(x, e^i),$$

where $e^i = (\underbrace{0, \ldots, 0}_{i-1}, 1, \underbrace{0, \ldots, 0}_{n-i})$.

- Hence, $x_i = \langle x, e^i \rangle_2 = \mathsf{A}(f(x), e^i)$

## The perfect case $\alpha(x) = 1$

- For every $i \in [n]$ it holds that

$$A(f(x), e^i) = b(x, e^i),$$

where $e^i = (\underbrace{0, \ldots, 0}_{i-1}, 1, \underbrace{0, \ldots, 0}_{n-i})$.

- Hence, $x_i = \langle x, e^i \rangle_2 = A(f(x), e^i)$
- Let $B(f(x)) = (A(f(x), e^1), \ldots, A(f(x), e^n))$

## Easy case: $\alpha(x) \geq 1 - \mathrm{neg}(n)$

### Fact 13

1. $b(x,w) \oplus b(x,y) = b(x, w \oplus y)$ *for every* $w, y \in \{0,1\}^n$.
2. $\forall r \in \{0,1\}^n$, *the rv* $(r \oplus R_n)$ *is uniformly distributed over* $\{0,1\}^n$ *(where* $R_n$ *is uniformly distributed over* $\{0,1\}^n$*).*

## Easy case: $\alpha(x) \geq 1 - \text{neg}(n)$

### Fact 13

1. $b(x, w) \oplus b(x, y) = b(x, w \oplus y)$ *for every* $w, y \in \{0, 1\}^n$.
2. $\forall r \in \{0, 1\}^n$, *the rv* $(r \oplus R_n)$ *is uniformly distributed over* $\{0, 1\}^n$ *(where* $R_n$ *is uniformly distributed over* $\{0, 1\}^n$*).*

Hence, $\forall i \in [n]$:

1. $x_i = b(x, r) \oplus b(x, r \oplus e^i)$ for every $r \in \{0, 1\}^n$

## Easy case: $\alpha(x) \geq 1 - \text{neg}(n)$

### Fact 13

1. $b(x, w) \oplus b(x, y) = b(x, w \oplus y)$ *for every* $w, y \in \{0,1\}^n$.
2. $\forall r \in \{0,1\}^n$, *the rv* $(r \oplus R_n)$ *is uniformly distributed over* $\{0,1\}^n$ *(where* $R_n$ *is uniformly distributed over* $\{0,1\}^n$*).*

Hence, $\forall i \in [n]$:

1. $x_i = b(x, r) \oplus b(x, r \oplus e^i)$ for every $r \in \{0,1\}^n$
2. $\Pr[\mathsf{A}(f(x), R_n) = b(x, R_n) \wedge \mathsf{A}(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)]$
   $\geq 1 - \text{neg}(n)$

## Easy case: $\alpha(x) \geq 1 - \mathrm{neg}(n)$

**Fact 13**

1. $b(x,w) \oplus b(x,y) = b(x, w \oplus y)$ *for every* $w, y \in \{0,1\}^n$.
2. $\forall r \in \{0,1\}^n$, *the rv* $(r \oplus R_n)$ *is uniformly distributed over* $\{0,1\}^n$ *(where* $R_n$ *is uniformly distributed over* $\{0,1\}^n$*).*

Hence, $\forall i \in [n]$:

1. $x_i = b(x,r) \oplus b(x, r \oplus e^i)$ for every $r \in \{0,1\}^n$
2. $\Pr[\mathsf{A}(f(x), R_n) = b(x, R_n) \wedge \mathsf{A}(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)] \geq 1 - \mathrm{neg}(n)$

# Easy case: $\alpha(x) \geq 1 - \mathrm{neg}(n)$

## Fact 13

1. $b(x, w) \oplus b(x, y) = b(x, w \oplus y)$ *for every* $w, y \in \{0,1\}^n$.
2. $\forall r \in \{0,1\}^n$, *the rv* $(r \oplus R_n)$ *is uniformly distributed over* $\{0,1\}^n$ *(where* $R_n$ *is uniformly distributed over* $\{0,1\}^n$*).*

Hence, $\forall i \in [n]$:

1. $x_i = b(x, r) \oplus b(x, r \oplus e^i)$ for every $r \in \{0,1\}^n$
2. $\Pr[\mathsf{A}(f(x), R_n) = b(x, R_n) \wedge \mathsf{A}(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)] \geq 1 - \mathrm{neg}(n)$

We let $\mathsf{B}(f(x)) =$
$(\mathsf{A}(f(x), R_n) \oplus \mathsf{A}(f(x), R_n \oplus e^1)), \ldots, \mathsf{A}(f(x), R_n) \oplus \mathsf{A}(f(x), R_n \oplus e^n))$.

## Intermediate case: $\alpha(x) \geq \frac{3}{4} + \frac{1}{q(n)}$

## Intermediate case: $\alpha(x) \geq \frac{3}{4} + \frac{1}{q(n)}$

For any $i \in [n]$

$$\begin{aligned} & \Pr[A(f(x), R_n) \oplus A(f(x), R_n \oplus e^i) = x_i] \qquad (4) \\ \geq \quad & \Pr[A(f(x), R_n) = b(x, R_n) \wedge A(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)] \end{aligned}$$

**Intermediate case:** $\alpha(x) \geq \frac{3}{4} + \frac{1}{q(n)}$

For any $i \in [n]$

$$\begin{aligned} & \Pr[A(f(x), R_n) \oplus A(f(x), R_n \oplus e^i) = x_i] \\ \geq \quad & \Pr[A(f(x), R_n) = b(x, R_n) \wedge A(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)] \\ \geq \quad & \frac{1}{2} + \frac{2}{q(n)} \end{aligned} \tag{4}$$

## Intermediate case: $\alpha(x) \geq \frac{3}{4} + \frac{1}{q(n)}$

For any $i \in [n]$

$$\begin{aligned}
&\Pr[A(f(x), R_n) \oplus A(f(x), R_n \oplus e^i) = x_i] \\
&\geq \Pr[A(f(x), R_n) = b(x, R_n) \wedge A(f(x), R_n \oplus e^i) = b(x, R_n \oplus e^i)] \\
&\geq \frac{1}{2} + \frac{2}{q(n)}
\end{aligned} \tag{4}$$

**Algorithm 14 (B)**

Input: $f(x) \in \{0,1\}^n$

1. For every $i \in [n]$
    1. Sample $r^1, \ldots, r^v \in \{0,1\}^n$ uniformly at random
    2. Let $m_i = \text{maj}_{j \in [v]}\{(A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i)\}$
2. Output $(m_1, \ldots, m_n)$

## B's success provability

The following holds for “large enough" $v = v(n)$.

**Claim 15**

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \text{neg}(n)$.

## B's success provability

The following holds for "large enough" $v = v(n)$.

**Claim 15**

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \mathrm{neg}(n)$.

Proof: For $j \in [v]$, let the indicator rv $W^j$ be $1$, iif $A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i) = x_i$.

## B's success provability

The following holds for "large enough" $v = v(n)$.

**Claim 15**

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \mathrm{neg}(n)$.

Proof: For $j \in [v]$, let the indicator rv $W^j$ be $1$, iif
$A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i) = x_i$.
We want to lowerbound $\Pr\left[\sum_{j=1}^{v} W^j > \frac{v}{2}\right]$.

## B's success provability

The following holds for “large enough" $v = v(n)$.

**Claim 15**

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \operatorname{neg}(n)$.

Proof: For $j \in [v]$, let the indicator rv $W^j$ be $1$, iif $A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i) = x_i$.

We want to lowerbound $\Pr\left[\sum_{j=1}^{v} W^j > \frac{v}{2}\right]$.

- The $W^j$ are iids and $E[W^j] \geq \frac{1}{2} + \frac{2}{q(n)}$ for every $j \in [v]$

## B's success provability

The following holds for "large enough" $v = v(n)$.

**Claim 15**

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \text{neg}(n)$.

Proof: For $j \in [v]$, let the indicator rv $W^j$ be $1$, iif $A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i) = x_i$.

We want to lowerbound $\Pr\left[\sum_{j=1}^{v} W^j > \frac{v}{2}\right]$.

- The $W^j$ are iids and $\text{E}[W^j] \geq \frac{1}{2} + \frac{2}{q(n)}$ for every $j \in [v]$

**Lemma 16 (Hoeffding's inequality)**

*Let $X^1, \ldots, X^v$ be iids over $[0,1]$ with expectation $\mu$. Then,* $\Pr\left[\left|\frac{\sum_{j=i}^{v} X^j}{v} - \mu\right| \geq \varepsilon\right] \leq 2 \cdot \exp(-2\varepsilon^2 v)$ *for every $\varepsilon > 0$.*

## B's success provability

The following holds for "large enough" $v = v(n)$.

### Claim 15

For every $i \in [n]$, it holds that $\Pr[m_i = x_i] \geq 1 - \text{neg}(n)$.

Proof: For $j \in [v]$, let the indicator rv $W^j$ be $1$, iif
$A(f(x), r^j) \oplus A(f(x), r^j \oplus e^i) = x_i$.
We want to lowerbound $\Pr\left[\sum_{j=1}^{v} W^j > \frac{v}{2}\right]$.

- The $W^j$ are iids and $E[W^j] \geq \frac{1}{2} + \frac{2}{q(n)}$ for every $j \in [v]$

### Lemma 16 (Hoeffding's inequality)

*Let $X^1, \ldots, X^v$ be iids over $[0,1]$ with expectation $\mu$. Then,*
$\Pr\left[\left|\frac{\sum_{j=i}^{v} X^j}{v} - \mu\right| \geq \varepsilon\right] \leq 2 \cdot \exp(-2\varepsilon^2 v)$ *for every $\varepsilon > 0$.*

We complete the proof taking $X^j = W^j$, $\varepsilon = 1/4q(n)$ and
$v \in \omega(\log(n) \cdot q(n)^2)$.

## The actual case: $\alpha(x) \geq \frac{1}{2} + \frac{1}{q(n)}$

## The actual case: $\alpha(x) \geq \frac{1}{2} + \frac{1}{q(n)}$

- What goes wrong?

## The actual case: $\alpha(x) \geq \frac{1}{2} + \frac{1}{q(n)}$

- What goes wrong?
- Idea: guess the values of $\{b(x, r^1), \ldots, b(x, r^v)\}$ (instead of calling $\{\mathsf{A}(f(x), r^1), \ldots, \mathsf{A}(f(x), r^v)\}$)

## The actual case: $\alpha(x) \geq \frac{1}{2} + \frac{1}{q(n)}$

- What goes wrong?
- Idea: guess the values of $\{b(x, r^1), \ldots, b(x, r^v)\}$ (instead of calling $\{\mathsf{A}(f(x), r^1), \ldots, \mathsf{A}(f(x), r^v)\}$)

  Problem: negligible success probability

## The actual case: $\alpha(x) \geq \frac{1}{2} + \frac{1}{q(n)}$

- What goes wrong?
- Idea: guess the values of $\{b(x, r^1), \ldots, b(x, r^v)\}$ (instead of calling $\{\mathsf{A}(f(x), r^1), \ldots, \mathsf{A}(f(x), r^v)\}$)

  Problem: negligible success probability

  Solution: choose the samples in a correlated manner

# Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^{\ell} - 1$.

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^{\ell} - 1$.

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^{\ell} - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^{\ell} - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^\ell - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

**Algorithm 17 (B)**

Input: $f(x) \in \{0,1\}^n$

1. Sample uniformly (and independently) $t_1, \ldots, t_\ell \in \{0,1\}^n$
2. Guess $\{b(x, t^i)\}_{i \in [\ell]}$
3. For all $\mathcal{L} \subseteq [\ell]$:
   Compute $b(x, r^{\mathcal{L}}) = \bigoplus_{i \in \mathcal{L}} b(x, t^i)$, where $r^{\mathcal{L}} := \bigoplus_{i \in \mathcal{L}} t^i$
4. For all $i \in [n]$, let $m_i = \mathrm{maj}_{\mathcal{L} \subseteq [\ell]}\{\mathrm{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}})\}$
5. Output $(m_1, \ldots, m_n)$

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^\ell - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

**Algorithm 17 (B)**

Input: $f(x) \in \{0,1\}^n$

1. Sample uniformly (and independently) $t_1, \ldots, t_\ell \in \{0,1\}^n$
2. Guess $\{b(x, t^i)\}_{i \in [\ell]}$
3. For all $\mathcal{L} \subseteq [\ell]$:
   Compute $b(x, r^{\mathcal{L}}) = \bigoplus_{i \in \mathcal{L}} b(x, t^i)$, where $r^{\mathcal{L}} := \bigoplus_{i \in \mathcal{L}} t^i$
4. For all $i \in [n]$, let $m_i = \mathrm{maj}_{\mathcal{L} \subseteq [\ell]} \{\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}})\}$
5. Output $(m_1, \ldots, m_n)$

- Fix $i \in [n]$, and let $W^{\mathcal{L}}$ be 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$.

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^\ell - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

**Algorithm 17 (B)**

Input: $f(x) \in \{0,1\}^n$

1. Sample uniformly (and independently) $t_1, \ldots, t_\ell \in \{0,1\}^n$
2. Guess $\{b(x, t^i)\}_{i \in [\ell]}$
3. For all $\mathcal{L} \subseteq [\ell]$:
   Compute $b(x, r^{\mathcal{L}}) = \bigoplus_{i \in \mathcal{L}} b(x, t^i)$, where $r^{\mathcal{L}} := \bigoplus_{i \in \mathcal{L}} t^i$
4. For all $i \in [n]$, let $m_i = \mathrm{maj}_{\mathcal{L} \subseteq [\ell]}\{\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}})\}$
5. Output $(m_1, \ldots, m_n)$

- Fix $i \in [n]$, and let $W^{\mathcal{L}}$ be 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$.
- We want to lowerbound $\Pr\left[\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}} > \frac{v}{2}\right]$

## Algorithm B

- Fix $\ell = \ell(n)$ (will be $O(\log n)$) and set $v = 2^\ell - 1$.
- In the following $\mathcal{L} \subseteq [\ell]$ stands for a non empty choice, and

**Algorithm 17 (B)**

Input: $f(x) \in \{0,1\}^n$

1. Sample uniformly (and independently) $t_1, \ldots, t_\ell \in \{0,1\}^n$
2. Guess $\{b(x, t^i)\}_{i \in [\ell]}$
3. For all $\mathcal{L} \subseteq [\ell]$:
   Compute $b(x, r^{\mathcal{L}}) = \bigoplus_{i \in \mathcal{L}} b(x, t^i)$, where $r^{\mathcal{L}} := \bigoplus_{i \in \mathcal{L}} t^i$
4. For all $i \in [n]$, let $m_i = \mathrm{maj}_{\mathcal{L} \subseteq [\ell]}\{\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}})\}$
5. Output $(m_1, \ldots, m_n)$

- Fix $i \in [n]$, and let $W^{\mathcal{L}}$ be 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$.
- We want to lowerbound $\Pr\left[\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}} > \frac{v}{2}\right]$
- Problem: the $W^{\mathcal{L}}$'s are dependent!

# Analyzing B's success probability

1. Let $T^1, \ldots, T^\ell$ be iid over $\{0,1\}^n$.
2. For every $\mathcal{L} \subseteq [\ell]$, let $R^{\mathcal{L}} = \bigoplus_{i \in \mathcal{L}} T^i$ .

## Analyzing B's success probability

1. Let $T^1, \ldots, T^\ell$ be iid over $\{0,1\}^n$.
2. For every $\mathcal{L} \subseteq [\ell]$, let $R^{\mathcal{L}} = \bigoplus_{i \in \mathcal{L}} T^i$ .

**Claim 18**

1. $\forall \mathcal{L} \subseteq [\ell]$, $R^{\mathcal{L}}$ is uniformly distributed over $\{0,1\}^n$
2. $\forall w, w' \in \{0,1\}^n$ and $\forall \mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$, it holds that $\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] = \Pr[R^{\mathcal{L}} = w] \cdot \Pr[R^{\mathcal{L}'} = w']$

## Analyzing B's success probability

1. Let $T^1, \ldots, T^\ell$ be iid over $\{0,1\}^n$.
2. For every $\mathcal{L} \subseteq [\ell]$, let $R^{\mathcal{L}} = \bigoplus_{i \in \mathcal{L}} T^i$ .

**Claim 18**

1. $\forall \mathcal{L} \subseteq [\ell]$, $R^{\mathcal{L}}$ is uniformly distributed over $\{0,1\}^n$
2. $\forall w, w' \in \{0,1\}^n$ and $\forall \mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$, it holds that $\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] = \Pr[R^{\mathcal{L}} = w] \cdot \Pr[R^{\mathcal{L}'} = w']$

- Proof?

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

$$\begin{aligned}
&\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] \\
&= \sum_{(t^2,\ldots,t^\ell)\in\{0,1\}^{(\ell-1)n}} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot \\
&\qquad \Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)]
\end{aligned}$$

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

$$
\begin{aligned}
&\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] \\
&= \sum_{(t^2,\ldots,t^\ell)\in\{0,1\}^{(\ell-1)n}} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot \\
&\qquad \Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i) = w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&\qquad \cdot \Pr[R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)]
\end{aligned}
$$

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

$$\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w']$$

$$= \sum_{(t^2,\ldots,t^\ell)\in\{0,1\}^{(\ell-1)n}} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot$$

$$\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)]$$

$$= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)]$$

$$\cdot \Pr[R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)]$$

$$= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot 2^{-n}$$

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

$$\begin{aligned}
&\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] \\
&= \sum_{(t^2,\ldots,t^\ell)\in\{0,1\}^{(\ell-1)n}} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot \\
&\qquad \Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&\qquad \cdot \Pr[R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot 2^{-n} \\
&= 2^{-n} \cdot 2^{-n}
\end{aligned}$$

## Proving Fact 18(2)

Assume wlg. that $1 \in (\mathcal{L}' \setminus \mathcal{L})$.

$$
\begin{aligned}
&\Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w'] \\
&= \sum_{(t^2,\ldots,t^\ell)\in\{0,1\}^{(\ell-1)n}} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot \\
&\qquad \Pr[R^{\mathcal{L}} = w \wedge R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&\qquad \cdot \Pr[R^{\mathcal{L}'} = w' \mid (T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \\
&= \sum_{(t^2,\ldots,t^\ell)\colon (\bigoplus_{i\in\mathcal{L}} t^i)=w} \Pr[(T^2,\ldots,T^\ell) = (t^2,\ldots,t^\ell)] \cdot 2^{-n} \\
&= 2^{-n} \cdot 2^{-n} \\
&= \Pr[R^{\mathcal{L}} = w] \cdot \Pr[R^{\mathcal{L}'} = w']
\end{aligned}
$$

□

## Pairwise independence variables

### Definition 19 (pairwise independent random variables)

A sequence of random variables $X^1, \ldots, X^v$ is pairwise independent, if $\forall i \neq j \in [v]$ and $\forall a, b$, it holds that

$$\Pr[X^i = a \wedge X^j = b] = \Pr[X^i = a] \cdot \Pr[X^j = b]$$

**Definition 19 (pairwise independent random variables)**

A sequence of random variables $X^1, \ldots, X^v$ is pairwise independent, if $\forall i \neq j \in [v]$ and $\forall a, b$, it holds that

$$\Pr[X^i = a \wedge X^j = b] = \Pr[X^i = a] \cdot \Pr[X^j = b]$$

- By Claim 18, $r^{\mathcal{L}}$ and $r^{\mathcal{L}'}$ (chosen by B) are pairwise independent for every $\mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$.

## Pairwise independence variables

### Definition 19 (pairwise independent random variables)

A sequence of random variables $X^1, \ldots, X^v$ is pairwise independent, if $\forall i \neq j \in [v]$ and $\forall a, b$, it holds that

$$\Pr[X^i = a \wedge X^j = b] = \Pr[X^i = a] \cdot \Pr[X^j = b]$$

- By Claim 18, $r^{\mathcal{L}}$ and $r^{\mathcal{L}'}$ (chosen by B) are pairwise independent for every $\mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$.
- Hence, also $W^{\mathcal{L}}$ and $W^{\mathcal{L}'}$ why?. (Recall, $W^{\mathcal{L}}$ is 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$)

# Pairwise independence variables

**Definition 19 (pairwise independent random variables)**

A sequence of random variables $X^1, \ldots, X^v$ is pairwise independent, if $\forall i \neq j \in [v]$ and $\forall a, b$, it holds that

$$\Pr[X^i = a \wedge X^j = b] = \Pr[X^i = a] \cdot \Pr[X^j = b]$$

- By Claim 18, $r^{\mathcal{L}}$ and $r^{\mathcal{L}'}$ (chosen by B) are pairwise independent for every $\mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$.
- Hence, also $W^{\mathcal{L}}$ and $W^{\mathcal{L}'}$ why?. (Recall, $W^{\mathcal{L}}$ is 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$)

# Pairwise independence variables

### Definition 19 (pairwise independent random variables)

A sequence of random variables $X^1, \ldots, X^v$ is pairwise independent, if $\forall i \neq j \in [v]$ and $\forall a, b$, it holds that

$$\Pr[X^i = a \wedge X^j = b] = \Pr[X^i = a] \cdot \Pr[X^j = b]$$

- By Claim 18, $r^{\mathcal{L}}$ and $r^{\mathcal{L}'}$ (chosen by B) are pairwise independent for every $\mathcal{L} \neq \mathcal{L}' \subseteq [\ell]$.
- Hence, also $W^{\mathcal{L}}$ and $W^{\mathcal{L}'}$ why?. (Recall, $W^{\mathcal{L}}$ is 1 iff $\mathsf{A}(f(x), r^{\mathcal{L}} \oplus e^i) \oplus b(x, r^{\mathcal{L}}) = x_i$)

### Lemma 20 (Chebyshev's inequality)

*Let $X^1, \ldots, X^v$ be pairwise-independent random variables with expectation $\mu$ and variance $\sigma^2$. Then, for every $\varepsilon > 0$,*

$$\Pr\left[\left|\frac{\sum_{j=1}^{v} X^j}{v} - \mu\right| \geq \varepsilon\right] \leq \frac{\sigma^2}{\varepsilon^2 v}$$

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$

# B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$
- Taking $\varepsilon = 1/2q(n)$ and $v = 2n/\varepsilon^2$ (i.e., $\ell = \lceil \log(2n/\varepsilon^2) \rceil$), yields that

$$\Pr[m_i = x_i] = \Pr\left[\frac{\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}}}{v} > \frac{1}{2}\right] \geq 1 - \frac{1}{2n} \tag{5}$$

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$
- Taking $\varepsilon = 1/2q(n)$ and $v = 2n/\varepsilon^2$ (i.e., $\ell = \lceil \log(2n/\varepsilon^2) \rceil$), yields that

$$\Pr[m_i = x_i] = \Pr\left[\frac{\sum_{\mathcal{L}\subseteq[\ell]} W^{\mathcal{L}}}{v} > \frac{1}{2}\right] \geq 1 - \frac{1}{2n} \tag{5}$$

## B's success provability cont

- Assuming that B always guesses $\{b(x,t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$
- Taking $\varepsilon = 1/2q(n)$ and $v = 2n/\varepsilon^2$ (i.e., $\ell = \lceil \log(2n/\varepsilon^2) \rceil$), yields that

$$\Pr[m_i = x_i] = \Pr\left[\frac{\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}}}{v} > \frac{1}{2}\right] \geq 1 - \frac{1}{2n} \tag{5}$$

- Hence, by a union bound, B outputs $x$ with probability $\frac{1}{2}$.

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$
- Taking $\varepsilon = 1/2q(n)$ and $v = 2n/\varepsilon^2$ (i.e., $\ell = \lceil \log(2n/\varepsilon^2) \rceil$), yields that

$$\Pr[m_i = x_i] = \Pr\left[\frac{\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}}}{v} > \frac{1}{2}\right] \geq 1 - \frac{1}{2n} \qquad (5)$$

- Hence, by a union bound, B outputs $x$ with probability $\frac{1}{2}$.

## B's success provability cont

- Assuming that B always guesses $\{b(x, t^i)\}$ correctly, then for every $\mathcal{L} \subseteq [\ell]$
  - $\mathrm{E}[W^{\mathcal{L}}] \geq \frac{1}{2} + \frac{1}{q(n)}$
  - $\mathrm{Var}(W^{\mathcal{L}}) := \mathrm{E}[W^{\mathcal{L}}]^2 - \mathrm{E}[(W^{\mathcal{L}})^2] \leq 1$
- Taking $\varepsilon = 1/2q(n)$ and $v = 2n/\varepsilon^2$ (i.e., $\ell = \lceil \log(2n/\varepsilon^2) \rceil$), yields that

$$\Pr[m_i = x_i] = \Pr\left[\frac{\sum_{\mathcal{L} \subseteq [\ell]} W^{\mathcal{L}}}{v} > \frac{1}{2}\right] \geq 1 - \frac{1}{2n} \tag{5}$$

- Hence, by a union bound, B outputs $x$ with probability $\frac{1}{2}$.
- Taking the guessing into account, yields that B outputs $x$ with probability at least $2^{-\ell}/2 \in \Omega(n/q(n)^2)$.

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ “pseudorandom bits"

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ "pseudorandom bits"
- Alternative proof for the LHL:
  Let $X$ be a rv with over $\{0,1\}^n$ with $\mathrm{H}_\infty(X) \geq t$, and assume $\mathrm{SD}((R_n, \langle R_n, X\rangle_2), (R_n, U_1)) > \alpha = 2^{-c \cdot t}$ for some universal $c > 0$.

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ “pseudorandom bits"
- Alternative proof for the LHL:
  Let $X$ be a rv with over $\{0,1\}^n$ with $H_\infty(X) \geq t$, and assume $SD((R_n, \langle R_n, X\rangle_2),(R_n, U_1)) > \alpha = 2^{-c \cdot t}$ for some universal $c > 0$.

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ “pseudorandom bits"
- Alternative proof for the LHL:
  Let $X$ be a rv with over $\{0,1\}^n$ with $\mathrm{H}_\infty(X) \geq t$, and assume $\mathrm{SD}((R_n, \langle R_n, X\rangle_2), (R_n, U_1)) > \alpha = 2^{-c \cdot t}$ for some universal $c > 0$.

$\implies$ Exists (a possibly inefficient) algorithm $D$ that distinguishes $(R_n, \langle R_n, X\rangle_2)$ from $(R_n, U_1)$ with advantage $\alpha$

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ “pseudorandom bits"
- Alternative proof for the LHL:
  Let $X$ be a rv with over $\{0,1\}^n$ with $\mathrm{H}_\infty(X) \geq t$, and assume $\mathrm{SD}((R_n, \langle R_n, X\rangle_2), (R_n, U_1)) > \alpha = 2^{-c \cdot t}$ for some universal $c > 0$.

$\implies$ Exists (a possibly inefficient) algorithm $D$ that distinguishes $(R_n, \langle R_n, X\rangle_2)$ from $(R_n, U_1)$ with advantage $\alpha$

$\implies$ Exists algorithm $\mathsf{A}$ that predicts $\langle R_n, X\rangle_2$ given $R_n$ with prob $\frac{1}{2} + \alpha$

# Reflections

- Hardcore functions:
  Similar ideas allows to output $\log n$ “pseudorandom bits"
- Alternative proof for the LHL:
  Let $X$ be a rv with over $\{0,1\}^n$ with $\mathrm{H}_\infty(X) \geq t$, and assume $\mathrm{SD}((R_n, \langle R_n, X\rangle_2), (R_n, U_1)) > \alpha = 2^{-c \cdot t}$ for some universal $c > 0$.

$\Longrightarrow$ Exists (a possibly inefficient) algorithm $D$ that distinguishes $(R_n, \langle R_n, X\rangle_2)$ from $(R_n, U_1)$ with advantage $\alpha$

$\Longrightarrow$ Exists algorithm A that predicts $\langle R_n, X\rangle_2$ given $R_n$ with prob $\frac{1}{2} + \alpha$

$\Longrightarrow$ (by GL) Exists algorithm B that guesses $X$ from nothing, with prob $\alpha^{O(1)} > 2^{-t}$

# Reflections cont.

- List decoding:
  An encoder $C\colon \{0,1\}^n \mapsto \{0,1\}^m$ and a decoder $D$, such that the following holds for any $x \in \{0,1\}^n$ and $c$ of hamming distance $\frac{1}{2} - \delta$ from $C(x)$:

## Reflections cont.

- List decoding:
  An encoder $C\colon \{0,1\}^n \mapsto \{0,1\}^m$ and a decoder $D$, such that the following holds for any $x \in \{0,1\}^n$ and $c$ of hamming distance $\frac{1}{2} - \delta$ from $C(x)$:

## Reflections cont.

- List decoding:
  An encoder $C\colon \{0,1\}^n \mapsto \{0,1\}^m$ and a decoder $D$, such that the following holds for any $x \in \{0,1\}^n$ and $c$ of hamming distance $\frac{1}{2} - \delta$ from $C(x)$:
  $D(c, \delta)$ outputs a list of size at most $\mathrm{poly}(1/\delta)$ that whp. contains $x$

# Reflections cont.

- List decoding:
  An encoder $C\colon \{0,1\}^n \mapsto \{0,1\}^m$ and a decoder $D$, such that the following holds for any $x \in \{0,1\}^n$ and $c$ of hamming distance $\frac{1}{2} - \delta$ from $C(x)$:
  $D(c, \delta)$ outputs a list of size at most $\mathrm{poly}(1/\delta)$ that whp. contains $x$

  The code we used here is known as the Hadamard code
- LPN - learning parity with noise:
  Find $x$ given polynomially many samples of $\langle x, R_n \rangle_2 + N$, where $\Pr[N = 1] \leq \frac{1}{2} - \delta$.

# Reflections cont.

- List decoding:
  An encoder $C\colon \{0,1\}^n \mapsto \{0,1\}^m$ and a decoder $D$, such that the following holds for any $x \in \{0,1\}^n$ and $c$ of hamming distance $\frac{1}{2} - \delta$ from $C(x)$:
  $D(c,\delta)$ outputs a list of size at most $\mathrm{poly}(1/\delta)$ that whp. contains $x$

  The code we used here is known as the Hadamard code
- LPN - learning parity with noise:
  Find $x$ given polynomially many samples of $\langle x, R_n\rangle_2 + N$, where $\Pr[N=1] \leq \frac{1}{2} - \delta$.

  The difference comparing to Goldreich-Levin – no control over the $R_n$'s.